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A NOTE ON THE EXACT VARIANCE OF PRODUCTS1

bу

Leo A. Goodman

Columbia University

Mathematical Statistics

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A number of readers of [2] have written the author inquiring about the possibility of generalizing the results presented there.

It therefore seemed worthwhile to prepare the present brief note indicating how some of the results in [2] can be generalized.

Let  $x_1, x_2, \ldots, x_K$  be K random variables. Let us denote the expected value of  $x_i$  by  $E(x_i) = X_i$ , the variance of  $x_i$  by  $V_i$ , and the square of the coefficient of variation of  $x_i$  by  $V_i/X_i = G_i$ . (For the sake of simplicity, we assume that  $X_i \neq 0$ , although some of the results presented do not require this assumption.) We shall make use of the simple identity

(1) 
$$\prod_{i=1}^{K} x_i = \prod_{i=1}^{K} x_i \prod_{i=1}^{K} (\delta_i + 1) = \prod_{i=1}^{K} (\Delta_i + x_i)$$

where  $\delta_i = (x_i - X_i)/X_i$  and  $\triangle_i = (x_i - X_i)$  . If the  $x_i$ 

are mutually independent, we find using identity (1) that the variance of  $\prod_{i=1}^{K} x_i$  will be equal to

$$(2) \quad V\left(\prod_{i=1}^{K} x_{i}\right) = E\left\{\prod_{i=1}^{K} x_{i}^{2}\right\} - \prod_{i=1}^{K} x_{i}^{2} = \prod_{i=1}^{K} x_{i}^{2} \left[\prod_{i=1}^{K} (G_{i} + 1) - 1\right],$$

which can also be written as  $V \left( \prod_{i=1}^{K} x_i \right) = \prod_{i=1}^{K} \left( V_i + X_i^2 \right) - \prod_{i=1}^{K} X_i^2$ 

$$(3) = \sum_{i=1}^{K} x_{i}^{1} \int_{1}^{2} x_{i}^{2} + \sum_{i=1}^{K} v_{i}v_{i} \int_{1}^{K} x_{i}^{2} + \sum_{i=1}^{K} v_{i}v_{i}v_{i} \int_{1}^{K} x_{i}^{2} + \cdots + v_{1}v_{2} \cdots v_{K}$$

$$= \int_{1}^{K} x_{i}^{2} \left[ \sum_{i=1}^{K} c_{i}^{2} + \sum_{i=1}^{K} c_{i}c_{i}^{2} + \sum_{i=1}^{K} c_{i}^{2} + \cdots + c_{i}^{2} + \cdots +$$

where the summation,  $\sum_{i_1,i_2,\cdots,i_S}$ , is over all values of

 $i_1 \neq i_2 \neq i_3 \dots \neq i_S$  ranging over 1,2, ..., K , and where is the product over the K-S values different  $j \neq i_1, i_2, \dots, i_S$ 

from the S values  $i_1, i_2, \dots, i_S$ . Equation (3) here is a generalization of equations (2) and (15) in [2] and equation (a) in [7]; equation (2) here appeared earlier in [3] where it was used to study the case where the distribution of  $\prod_{i=1}^{K} x_i$  was (approximately) logarithmic-normal.

We now present an unbiased estimator of  $V(\prod_{i=1}^{n} x_i)$  based on unbiased estimators,  $\overline{x}_i$  and  $v_i$ , of  $X_i$  and  $V_i$ , respectively, where  $\overline{x}_i$  is the sample mean and  $v_i$  is the sample variance in a sample of  $n_i$  observations each having mean  $X_i$  and variance  $V_i$  (i=1,2,...,K). When the K samples (i=1,2,...,K) are mutually independent, we find that

$$(4) \ \mathbf{v}(\prod_{\mathbf{i}=\mathbf{1}}^{\mathbf{K}} \mathbf{z_i}) = \prod_{\mathbf{i}=\mathbf{1}}^{\mathbf{K}} (\mathbf{v_i} + \mathbf{z_i}) - \prod_{\mathbf{i}=\mathbf{1}}^{\mathbf{K}} \mathbf{z_i}$$

$$= \prod_{i=1}^{K} \left[ \overline{x}_{i}^{2} + v_{i}(n_{i}-1)/n_{i} \right] - \prod_{i=1}^{K} \left[ \overline{x}_{i}^{2} - v_{i}/n_{i} \right]$$

is an unbiased estimator of  $V(\prod_{i=1}^K x_i)$ , where  $z_i = \overline{x}_i^2 - v_i/n_i$ .

This follows from the fact that  $E(\bar{x}_i)-X = V_i/n_i$ . Equation (4) here is a generalization of equation (5) in [2].

The case where the  $x_i$  are not mutually independent is more complicated. From identity (1) we see that the variance of  $\prod_{i=1}^K x_i$  is

(5) 
$$V(\prod_{i=1}^{K} x_i) = \prod_{i=1}^{K} x_i^2 \left[ E \left\{ \prod_{i=1}^{K} (\delta_i + 1)^2 \right\} - B^2 \right]$$
$$= E \left\{ \prod_{i=1}^{K} (\Delta_i + x_i)^2 \right\} - M^2$$

where  $M = E \left\{ \begin{array}{c} \frac{K}{1-1} & x_1 \\ 1-1 & x_1 \end{array} \right\}$  and  $B=M/\prod_{i=1}^{K} X_i$ . The special case of

(5) where K=2 was studied in [2]. We now consider the case where K=3 . By straightforward calculation, we find that, when K=3 , equation (5) can be rewritten as

(6) 
$$V(\prod_{i=1}^{3} x_i) = \prod_{i=1}^{3} X_i^2 \left[ \frac{3}{1-1} G_i + (B-1)(3-B) + \sum_{j,k,\ell} E \left\{ \delta_j^j \delta_k^k \delta_\ell^\ell \right\} \right],$$

where the indices j, k,  $\ell$  range over the values 0,1,2, and where  $h(j,k,\ell)$  is a symmetric function of j,k,  $\ell$  having the following values:

$$h(j,k,\ell) = \begin{cases} 0 \text{ for } (j,k,\ell) = (0,0,0), (0,0,1), (0,1,1) \\ 1 \text{ for } (j,k,\ell) = (0,2,2), (2,2,2) \\ 2 \text{ for } (j,k,\ell) = (0,1,2), (1,2,2) \\ 4 \text{ for } (j,k,\ell) = (1,1,1), (1,1,2) \end{cases}$$

Equation (6) here is a generalization of equation (18) in [2], the formula given there for the variance of the product of two random variables (not necessarily independent). In the same way that equation (18) was used in [2] to derive other variance formulas for various product estimators (e.g., equations (20) and (21) in [2]), equation (6) here can also be used to derive other variance formulas for product estimators where, for example, three estimators (rather than two) are multiplied together. We shall not go into these details in this brief note.

## References

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